

# The Impact of Social Nudges on User-Generated Content for Social Network Platforms

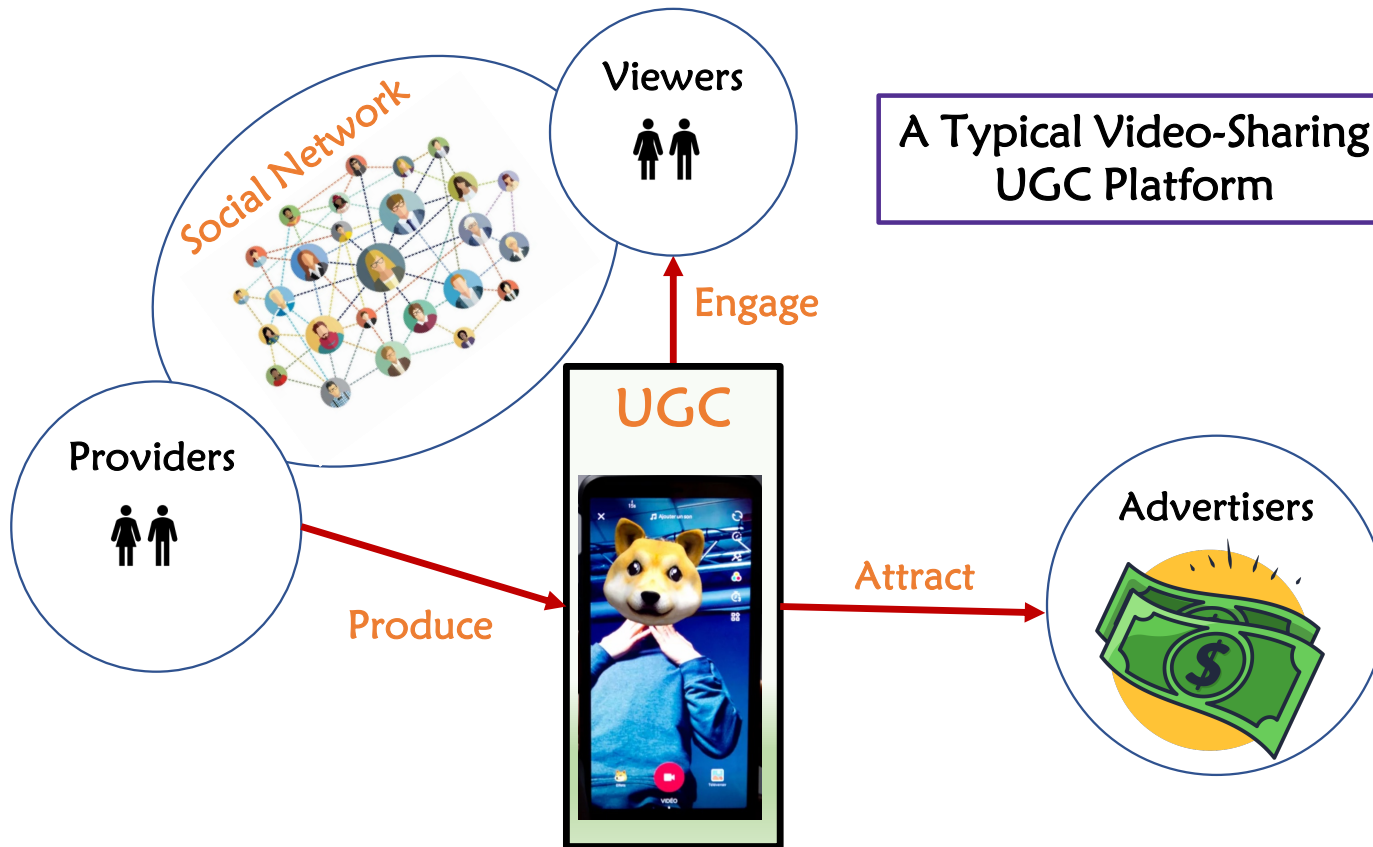
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Renyu (Philip) Zhang

(Joint works with Zhiyu Zeng, Hengchen Dai, Dennis J. Zhang, Heng Zhang, Max Shen)

- **Introduction and Contributions**
- Experiments and Empirics
- Social Network Model and Global Effect

# Video-Sharing UGC Platforms



Boosting the production of UGC is crucial to the platform!

- The platform has **little power** over providers.
- Social interactions (e.g., **likes** and **comments**) not only engage viewers but also motivate providers to **produce more**.

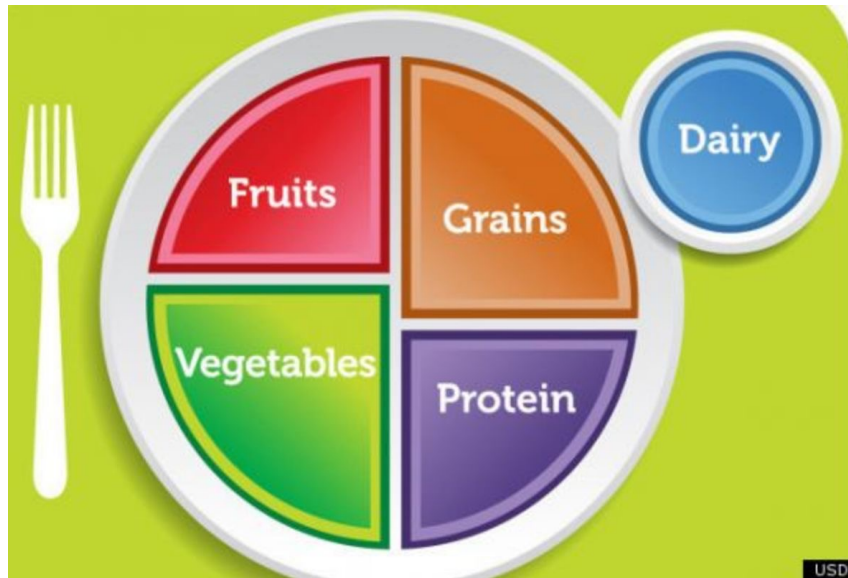
How can a UGC platform leverage the embedded **social network** to motivate **content production**?

# Nudge: Intervention without Money or Restriction

- **Nudge**: Intervention that intends to change individual behaviors **without altering financial incentives or imposing restrictions** (Thaler and Sunstein 2009).
- Nudge is a central concept in behavioral science and very widely applied in practice:



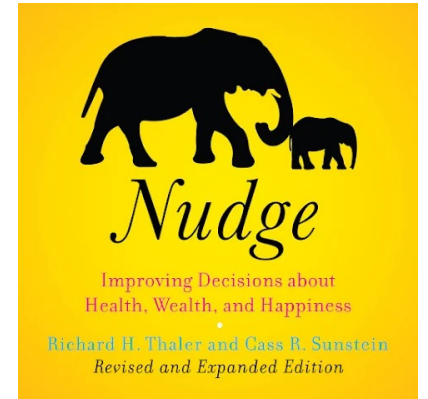
Photo credit: Xiaolong Li



- Nudge for healthy food:
- + 18% fruit and +25% veg consumptions
  - +15.3% healthier dietary or nutrition choices

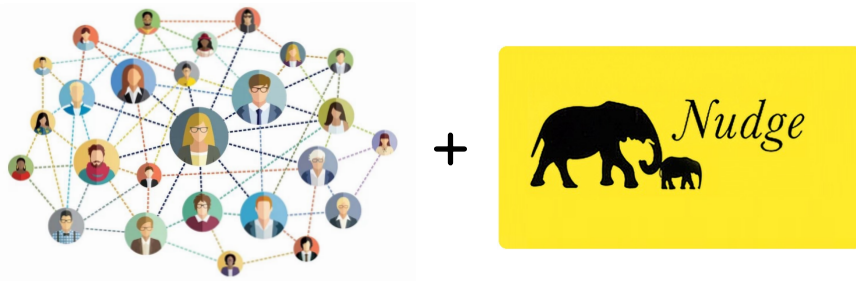
- Nudge for COVID-19 Vaccination:
- +84% appointment rate
  - +26% vaccination rate

Can a UGC platform use **nudge** to **motivate production**?

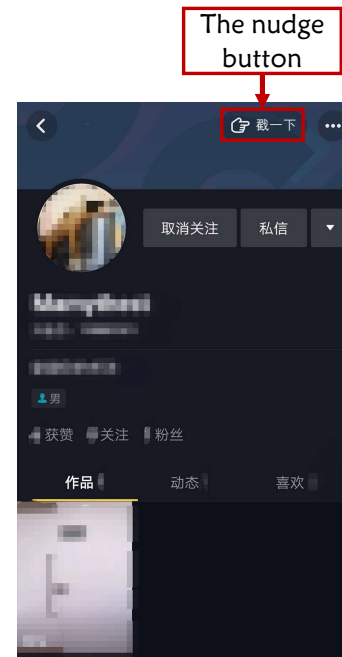


# Social Nudge


Social Nudge =

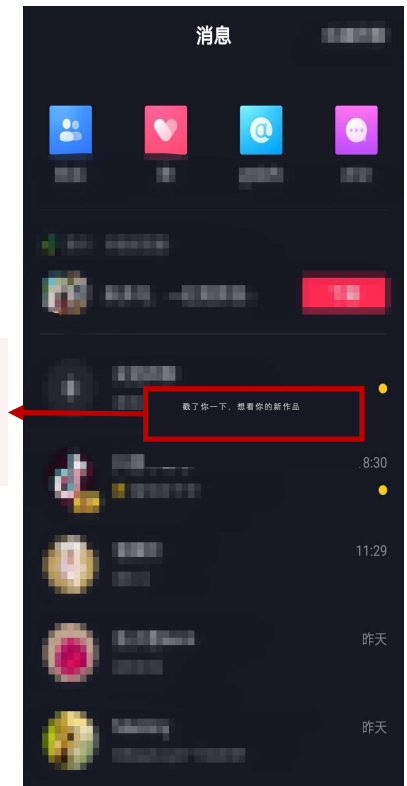


- **Social Nudge:** A user encourages a provider s/he follows to produce more: Conveying subtle peer recognition.
- **Literature:** Peer recognition may not work; not clear how social network plays a role.
- **Practice:** A low-cost (less than 1 week of work for an SDE) and novel product function.
- Different from likes and comments.



Screen Shot of the Profile Page of the provider you follow.

  
Philip poked you and wanted to see your new posts!



Screen Shot of the Message Box of the nudge recipient.

# Key Research (and Business) Questions

- How do social nudges affect the **UGC production of the recipients**?
  - **Literature**: Peer recognition NOT accompanied by financial incentives may **backfire**.
- How does the **underlying social network** impact social nudges?
- How to **quantify** the **global effect** of social nudges on the **entire social network**, considering **nudge diffusion**?

Combined Analytics Framework:  
**Field Experiment (A/B Testing) + Social Network Modeling**



## Related Literature

- **Content production on UGC platforms:** Peer recognition **NOT** accompanied by financial incentives may **backfire**.
  - Burtch et al. (2018), Huang et al. (2019), Restivo and Van de Rijt (2014), Gallus et al. (2020), etc.
- **Peer effects on social networks:** People's behaviors are largely influenced by those of their **neighbors on the network**.
  - Jackson (2010), Ballester et al. (2006), Candogan et al. (2012), Bimpkis et al. (2016), Zhou and Chen (2016), Gelper et al. (2021), etc.
- **Information-based interventions:** Offering the right information to the right people improves the **customer engagement** and **service capacity**.
  - Buell and Norton (2011), Buell et al. (2017), Parker et al. (2016), Cui et al. (2019), Xu et al. (2021), etc.
- **Platform operations:** Optimizing **operational strategies** for online platforms.
  - Banerjee et al. (2016), Cachon et al. (2017), Bai et al. (2019), Kabra et al. (2020), etc.



# Highlight of Main Contributions

- **Social nudge: A Novel Intervention to Boost UGC Production**

- **Direct effect:** Receiving social nudges **boosts the production** of the provider by **+13.21%**.
- **Low-cost, information-based** and **flexible** approach leveraging the embedded **social network** to address a central platform operations problem: Increasing UGC production.

- **Impact of the Social Network on Social Nudges**

- **Indirect effect (diffusion):** Nudge receivers send **+15.57%** more social nudges to providers.
- Both direct and indirect effects of social nudges are stronger with a **stronger tie** and last for **a few days**.

- **Global Effect of Social Nudges under Interferences**

- A **social network model** captures the direct and indirect effects of social nudges.
- **Bonacich Centrality for Edges** helps **quantify** the **global effect** of social nudges ( **$\sim +1\%$**  of total production).



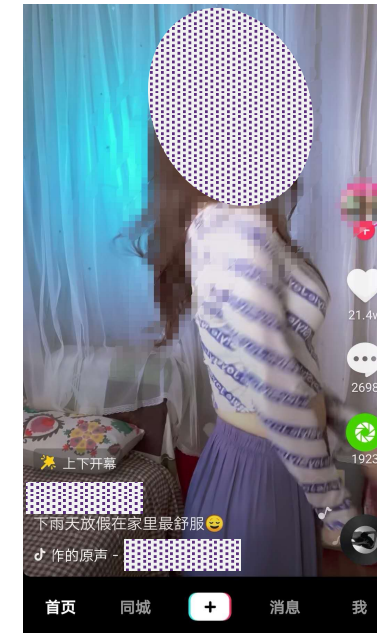
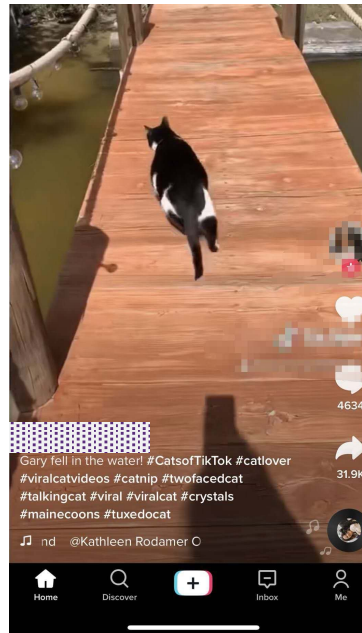


- Introduction and Contributions
- **Experiments and Empirics**
- Social Network Model and Global Effect

## Field Setting

- A Chinese online short-video sharing social network platform (referred to as **Platform O** hereafter).
- **300 million** of daily-active users (DAU), **half-billion** MAU, **20 million USD** advertising revenue per day.
- Each user can have two roles: Content provider and content viewer.
- Each user can **follow** other users: An embedded **social network** on Platform O.

Typical User-Interfaces of a Short-Video Sharing Platform

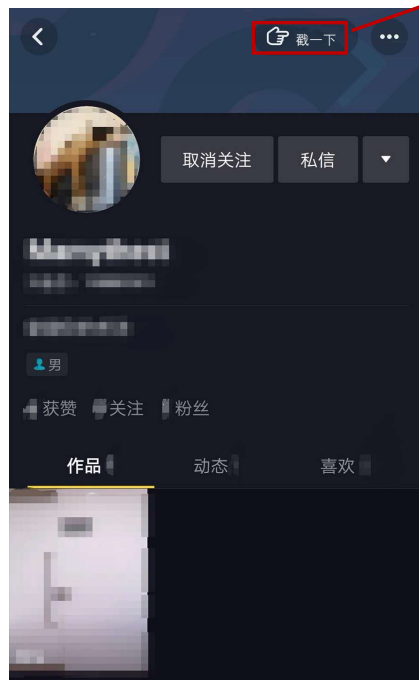


# Experiment Design

Followers sent social nudges to providers

The nudge button

Could see nudges sent by their followers



Sample of the Social Nudge Experiment



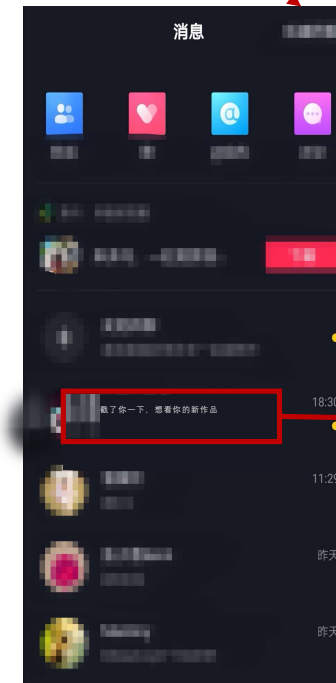
Providers were sent at least one nudge in the experiment



Treatment



Control



Philip poked you and wanted to see your new posts!

Could NOT see nudges sent by their followers

- Main experiment:
  - Sept. 12, 2018 ~ Sept. 14, 2018
  - Sample size: 993,676 providers (496,976 treatment and 496,700 control)
    - They had never received any social nudges before the experiment.
    - Most (94%) received only one social nudge during this experiment.
- Replication experiment (for robustness check):
  - Sept 14, 2018 ~ Sept. 20, 2018.
  - 655,001 providers non-overlapping with the main experiment.
- Data from both experiments pass the randomization checks, i.e., treatment and control providers are comparable.
- Outcome variables:
  - Number of videos uploaded.
  - Number of social nudges sent to other providers by the recipient.

**Question:** How do social nudges impact UGC production on a social network platform?

# Social Nudges Boost UGC Production

- We focus on the day a provider was sent the first social nudge in the experiment (**the first reception day**).
- Model specifications (a two-way tie means the nudge receiver also follows the sender):

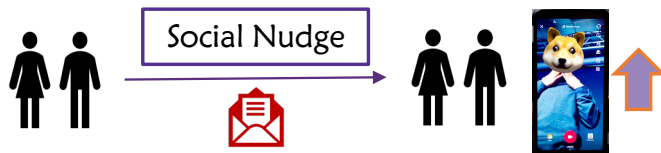
$$\text{Number of Videos Uploaded}_i = \beta_0 + \beta_1 \text{Treatment}_i + \epsilon_i$$

$$\text{Number of Videos Uploaded}_i = \beta_0 + \beta_1 \text{Treatment}_i + \beta_2 \text{Two-Way-Tie}_i + \alpha_3 \text{Treatment}_i * \text{Two-Way-Tie}_i + \epsilon_i$$

- Direct effect on production:

Treatment Effect	Average	2-way Tie	1-way Tie
Relative Effective Size	+13.21%***	+17.40%***	+9.35%***

Note: All p-values are smaller than 0.001.



- Social nudges **boost UGC production**.
- The direct effect of social nudges is stronger with a **stronger tie**.
- Social nudges increase the **video views, likes, and comments** of the providers; are more effective than **platform-initiated nudges**; and do **NOT** change video **quality (like-rate and comment-rate)**. The additional **likes** and **comments** after the experiment started are **NOT** the main reason for the UGC production boost.

# Social Nudges Diffuse

- How does the underlying **social network** impact the effect of social nudges?
- Model specifications:

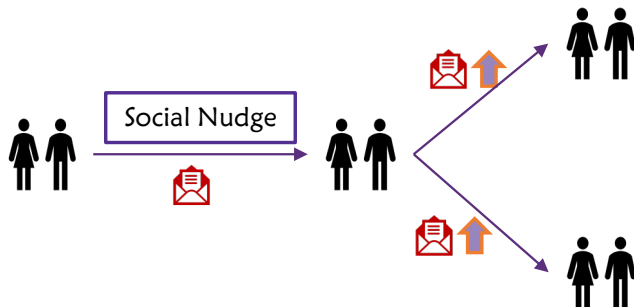
$$\text{Number of Social Nudges Sent}_i = \beta_0 + \beta_1 \text{Treatment}_i + \epsilon_i$$

$$\text{Number of Social Nudges Sent}_i = \beta_0 + \beta_1 \text{Treatment}_i + \beta_2 \text{Two-Way-Tie}_i + \alpha_3 \text{Treatment}_i * \text{Two-Way-Tie}_i + \epsilon_i$$

- Indirect effect of social nudges:

Treatment Effect	Average	2-way Tie	1-way Tie
Relative Effective Size	+15.57%***	+30.02%***	+2.87%*

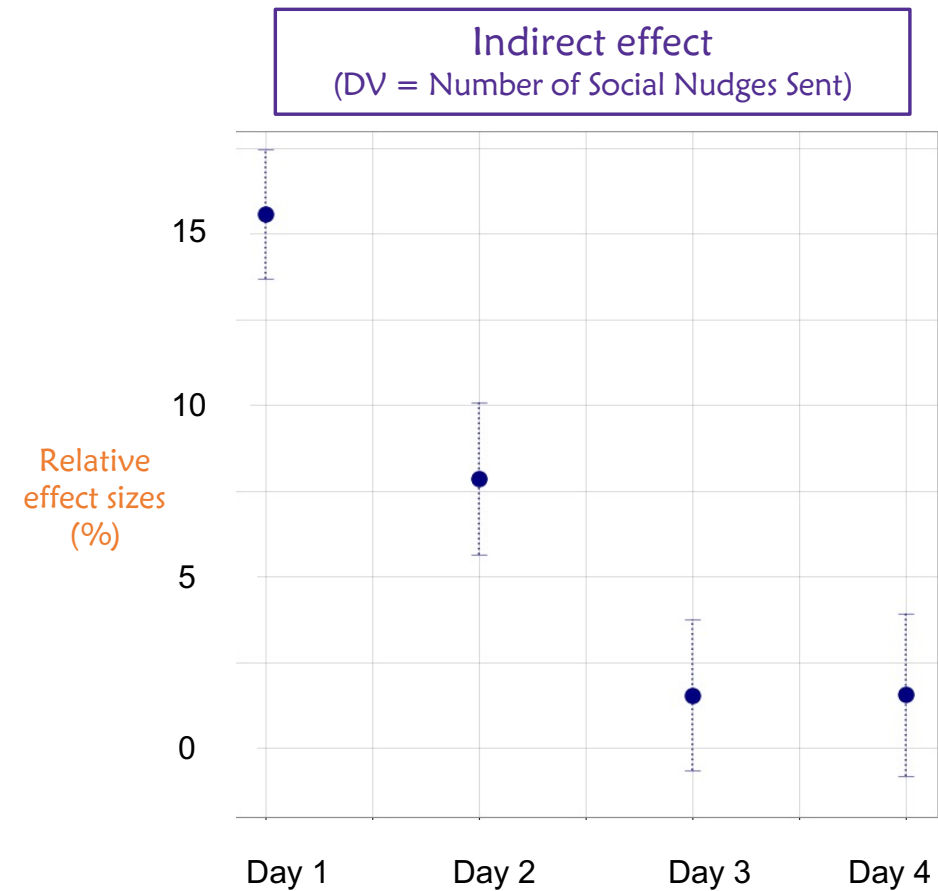
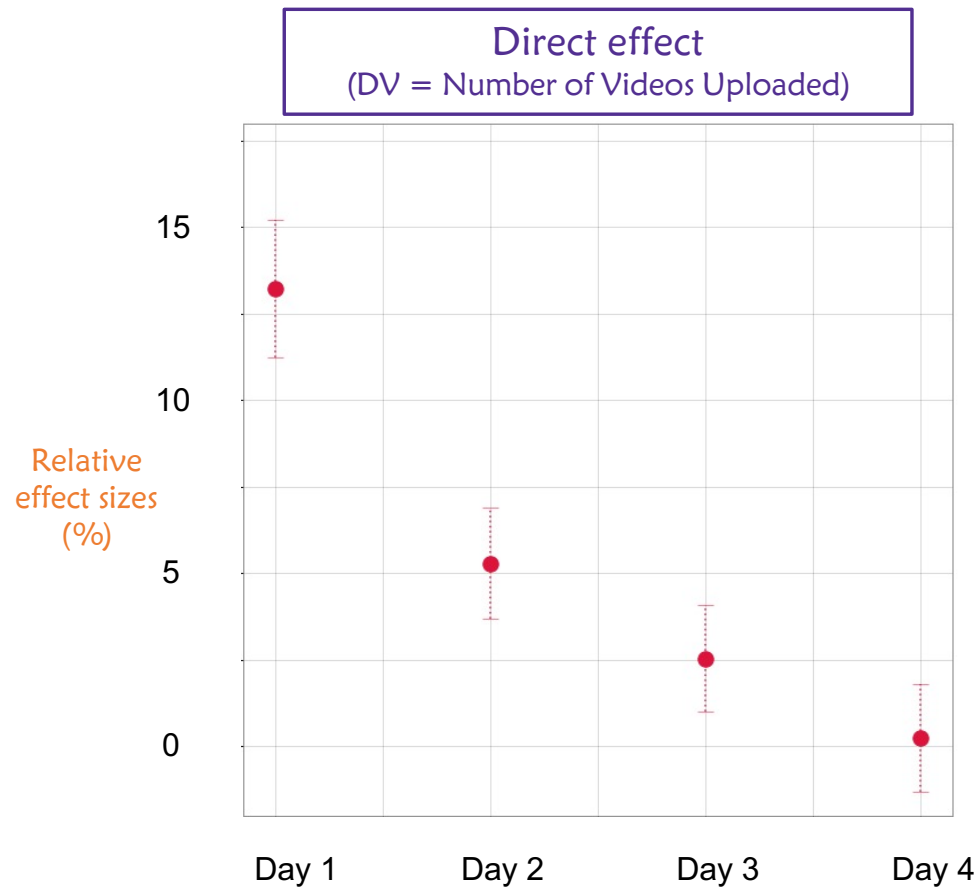
Note: \*\*\*: p-value<0.001; \*: p-value<0.05.



- Social nudges **diffuse on the social network** platform.
- The social nudge diffusion is stronger with a **stronger tie**.

# Over-Time Effects of Social Nudges

- How do the direct and indirect effects of social nudges **evolve over time**?



Note: The error bars are the 95% confidence intervals.



# Recap of the Social Nudge Experiment

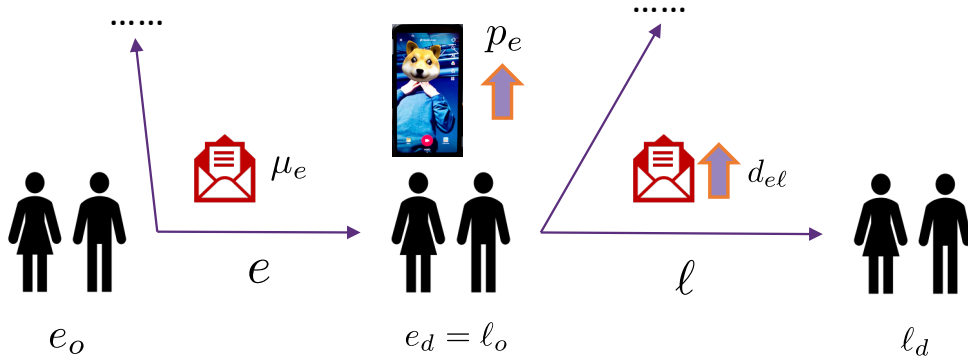


- How to evaluate the global effect of social nudges on the entire social network?
  - The global effect is NOT directly measurable by user-side randomized experiments due to the interference from nudge diffusion.

- Introduction and Contributions
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- **Social Network Model and Global Effect**

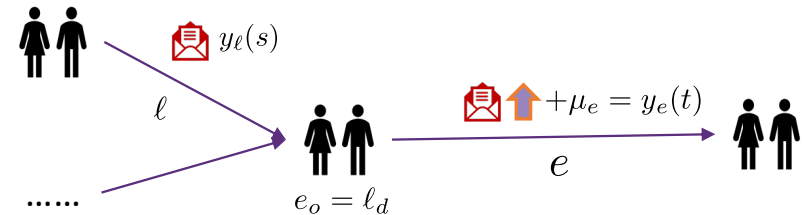
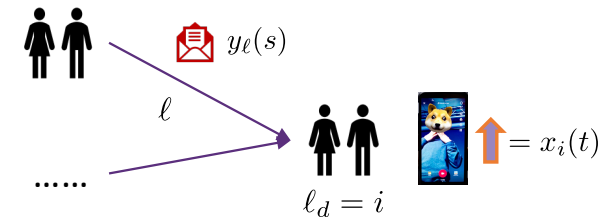
# Social Network Model

- A social network  $G = (V, E)$ , where  $V$  is the set of users and  $E$  is the set of following relationships.
- Discrete time period:  $t=1,2,3,\dots$  ( $t=1$  is the day when the social nudge function was launched).
- $\mu_e$ : the number of social nudges  $e_o$  sends to  $e_d$  without receiving any nudge (called the **organic nudges**).
- $p_e$ : the production boost of  $e_d$  after receiving a social nudge sent by  $e_o$ , i.e., the direct effect of social nudge.
- $d_{el}$ : the increase in the number of nudges sent on  $\ell$  after  $e_d$  receives a nudge sent by  $e_o$  ( $e_d = \ell_o$ ), i.e., the indirect effect.



Direct and indirect effects **decay to 0 exponentially** as  $p_e \alpha_p^{\Delta t}$  and  $d_{el} \alpha_d^{\Delta t}$ .

$$D := (d_{\ell e} : (\ell, e) \in E^2) \quad \mu := (\mu_e : e \in E) \quad \eta := (p_e / (1 - \alpha_p) : e \in E)$$



- The **production boost** of **user i** in period  $t$  (direct effect):

$$x_i(t) = \sum_{s=1}^t \alpha_p^{t-s} \sum_{\ell_d=i} p_{\ell} y_{\ell}(s) + \epsilon_i^x(t)$$

- The number of **social nudges** on **edge e** in period  $t$  (indirect effect):

$$y_e(t) = \mu_e + \sum_{s=1}^t \alpha_d^{t-s} \sum_{\ell_d=e_o} d_{\ell e} y_{\ell}(s) + \epsilon_e^y(t)$$

# Bonacich Centrality for Edges and the Equilibrium

- We care about the expected **equilibrium/stationary** state, i.e.,  $t$  goes to infinity.
- Equilibrium nudge vector  $\mathbf{y}^* := \lim_{t \uparrow +\infty} \mathbb{E}[\mathbf{y}(t)]$  and production boost  $\mathbf{x}^* := \lim_{t \uparrow +\infty} \sum_{i \in V} \mathbb{E}[x_i(t)]$ .

**Theorem.** The equilibrium nudge vector is  $\mathbf{y}^* = \mathcal{BE}(\mathbf{D}, \mu)$  and the equilibrium production boost is  $\mathbf{x}^* = \eta^T \mathbf{y}^*$ .

- Bonacich Centrality for Edges (BCE)** ( $\mathbf{I}$  is the identity matrix, assuming the limit exists):

$$\mathcal{BE}(\mathbf{D}, \mu) = \left( \mathbf{I} - \frac{1}{1 - \alpha_d} \mathbf{D} \right)^{-1} \mu = \underbrace{\left( \mathbf{I} \right)}_{\text{Direct Effect}} + \underbrace{\sum_{k=1}^{+\infty} \frac{1}{(1 - \alpha_d)^k} \mathbf{D}^k}_{\text{Indirect Effect}} \mu$$

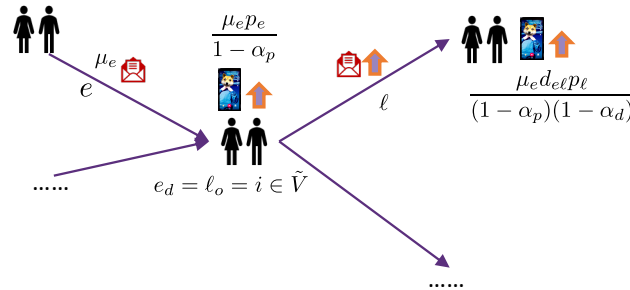
- The nudge vector of the  $k^{\text{th}}$ -order diffusion:  $\frac{1}{(1 - \alpha_d)^k} \mathbf{D}^k \mu$ ; the production boost from  $k^{\text{th}}$ -order diffusion:  $\frac{1}{(1 - \alpha_d)^k} \eta^T \mathbf{D}^k \mu$
- Core challenge: Dimension of the diffusion matrix  $\mathbf{D}$  for Platform O is  **$10^{32}$** .

# Approximation for the Global Effect

- Dimension of the diffusion matrix  $\mathbf{D}$  for Platform O:  $10^{32}$ .
  - First approximation:** Consider the **first-order diffusion** only and ignore the higher-order diffusions ( $\sim -0.72\%$  gap for Platform O).

$$\tilde{\mathbf{y}}(1) = \widetilde{\mathcal{B}\mathcal{E}}(\mathbf{D}, \mu, 1) = \left( \mathbf{I} + \underbrace{\frac{1}{1 - \alpha_d} \mathbf{D}}_{\text{First-order Diffusion}} \right) \mu, \quad \tilde{\mathbf{x}}(1) = \eta^T \tilde{\mathbf{y}}(1)$$

- Second approximation:** Randomly down-sample a subset of providers  $\tilde{V}$ , compute  $\tilde{x}(1, \tilde{V}) := \sum_{i \in \tilde{V}} \left( \sum_{e \in E, e_d = i} \underbrace{\left( \frac{\mu_e p_e}{1 - \alpha_p} \right)}_{\text{Direct Effect}} + \underbrace{\sum_{\ell \in E, e_d = \ell_o = i} \frac{\mu_e d_{e\ell} p_\ell}{(1 - \alpha_p)(1 - \alpha_d)}}_{\text{First-order Diffusion}} \right)$  and scale it to the entire social network by the inverse sampling rate.



**Proposition.**  $\kappa \cdot \tilde{x}(1, \tilde{V})$  is an **unbiased estimate** of  $\tilde{\mathbf{x}}(1)$ , where  $\kappa := |V|/|\tilde{V}|$  is the scaling factor.

# Parameter Estimation with Experimental Data

- To quantify the global effect of social nudges, it suffices to estimate  $(\mathbf{D}, \mathbf{p}, \mu, \alpha_p, \alpha_d)$  with data from the experiments.
- $\mu_e$  (the likelihood of sending organic nudges): Sampling edges in  $E$  whose **origin was in the control group** and cross-validate a logistic regression model.
- Jointly estimate  $(p_e, \alpha_p)$  by minimize the mean-squared-error between the **true over-time direct effect** (from the experiment) and the one predicted by our social network model  $p_e \alpha_p^t$  for  $t = 0, 1, 2$ .
- Jointly estimate  $(d_{el}, \alpha_d)$  by minimize the mean-squared-error between the **true over-time indirect effect** (from the experiment) and the one predicted by our social network model  $d_{el} \alpha_d^t$  for  $t = 0, 1$ .
- The estimates can be **replicated** by the data from **the second experiment** (relative gap **<10%** for each parameter), suggesting the **robustness** and **accuracy** of our approach.

The estimated parameter values are the **inputs of the social network model** to evaluate the global effect of **social nudges**,  $\kappa \cdot \tilde{x}(1, \tilde{V})$ .

# Estimation Results of the Global Effect

- **Baseline estimation:** Global Effect = **+48.65** videos per day
  - Number of providers who received any social nudges per day **multiplies** the production boost of receiving social nudges (estimated from the experiment).

Estimation with Our Social Network Model	
Global Effect = <b>+140.67</b> videos per day ( <b>~+1%</b> of the total UGC production of Platform O)	
Accumulated <b>Direct Effect</b> = <b>+130.08</b> videos per day	Accumulated <b>Indirect Effect</b> (diffusion) = <b>+10.59</b> videos per day
<b>167% higher</b> than the baseline estimation	Amounting to <b>22%</b> of the baseline estimation and <b>8.14%</b> of the direct effect

Note: The absolute global effect is scaled to protect sensitive data.

- The estimation of global effect is **robust** and **accurate**, which can be **replicated by the second experiment**.
  - The ratio between the **indirect effect** and the **direct effect** is **8.38%**.

**Insight:** Our **social network model** helps **de-bias** the **substantial underestimates** of the natural baseline estimation approach.



# Takeaways

- Develop the **social nudge** as a novel **low-cost**, **information-based**, and **flexible intervention** that leverages the embedded **social network** to **increase content production** on a large-scale UGC platform.
  - Social nudges **boost UGC production** and **diffuse** on the social network.
- **Social network model**: Evaluate the **global effect** of social nudges on a social network with the **interferences** from **nudge diffusion**.
- The **combined framework** of A/B testing and social network model has the potential to study other operations problems for **large-scale social network platforms**.

Link to the paper: [https://rphilipzhang.github.io/rphilipzhang/Social\\_nudge\\_nonblinded.pdf](https://rphilipzhang.github.io/rphilipzhang/Social_nudge_nonblinded.pdf)

# Thank You!

## Questions?

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<https://rphilipzhang.github.io/rphilipzhang/index.html>